On Iterative Neural Network Pruning, Reinitialization, and the Similarity of Masks

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In the context of the Lottery Ticket Hypothesis (Frankle & Carbin, 2018)... Do different pruning methods identify the same "winning ticket"? Or is there more than one?

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Research Findings

- 1. ∃ multiple different lottery tickets within an over-parametrized network
- 2. possible to find them through a variety of pruning techniques [extends the findings of Zhou et al. (2019)]
- 3. random structured pruning > random unstructured pruning
 - nets are more resistant to the removal of random units/ channels than random individual connections

[experiments shown here for LeNet on MNIST]



Effective Fraction of Pruned Weights

Research Findings

and similarly performing networks make different mistakes on held-out test sets

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	Number of examples in the MNIST test set over L2 structur								structured -	10000	3821	3221	3652	1555	2009	3676				
		W IO	hich t runino	he su i tech	b-netw nique a	/orks o aaree c	obtained on the pr ter 18 pr	ed t pre	through ea rediction, or runing itera	ch า	L ₁	structured -	3821	10000	3081	3760	1398	2024	3848	
		a\	/erage	e (ove	r 5 see	eds), af		pru		iterat	ions:	L_1 un	structured -	3221	3081	10000	3291	2570	3230	3276
hybrid - random structured - $L_{-\infty}$ structured -									3652	3760	3291	10000	1144	1514	7774					
									1555	1398	2570	1144	10000	8339	1127					
									2009	2024	3230	1514	8339		1466					
fc-only L_1 unstructured -										3676	3848	3276	7774	1127	1466	10000				
Sub-network accuracies at each pruning iteration + ensembling:										L ₂ structured -	L ₁ structured -	unstructured -	hybrid -	om structured-	-∞ structured -	unstructured -				
Pruning Iteration	L_2 S	L_1 S	L_1 US	hybrid	random S	$L_{-\infty}$ S	fc-only L_1	L_1 US	JS all	hybrid + fc-only L_1 US	US					L ₁		rando	L.	uly L ₁
18	36.7	37.8	32.4	81.8	11.3	14.5	87.4	4	91.0	91	.6									fc-01
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5. networks pruned via different iterative pruning techniques learn vastly different functions of their input,



techniques, even without late resetting

(x-axis), for weights in the 2nd convolutional layer in the LeNet architecture (seed: 0)



Should I rewind or fine-tune? What's the difference?

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Research Findings

4. patterns similar to structured pruning (~feature selection). Not true for finetuning.

LeNet conv1 weights





structured pruning + rewinding

AlexNet conv2 weights



unstructured pruning + rewinding

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lottery ticket-style weight rewinding, coupled with unstructured pruning, gives rise to connectivity

unstructured pruning + rewinding



unstructured pruning + finetuning

VGG11 conv2 weights



unstructured pruning + rewinding

Questions?

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